

# Note for 01/07/2021

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## 1 What to do

### 1.1 Plan for this week

1. Make three different distributions with distinct copula method, corresponding to three cell types.
2. Combine the three datasets as one, and plot a matrix of scatterplots
3. Check the shape of each plot if it resembles the cell data.
4. If not, try with different  $\mu$  and  $\Sigma$  for each copula to make a L shape, linear, and etc.

## 2 Copula

### 2.1 Definition of Copula

**Gaussian Copula** is a method to transform the random variable  $X = (X_1, \dots, X_p)$  to a new random variable  $f(X) = (f_1(X_1), \dots, f_p(X_p))$  with assumption that  $f(X)$  is multivariate Gaussian.

- A nonparametric extension of the normal, **nonparanormal**, depends on the functions  $f_j$ , and a mean  $\mu$  and covariance matrix  $\Sigma$
- When  $f_j$  are monotone and differentiable,  $NPN(\mu, \Sigma, f)$  is a Gaussian Copula.
- If  $X \sim NPN(\mu, \Sigma, f)$  is nonparanormal and each  $f_j$  is differentiable, then  $X_i \perp\!\!\!\perp X_j | X_{\setminus\{i,j\}}$  iff  $\Omega_{ij} = 0$  where  $\Omega = \Sigma^{-1}$

### 2.2 Covariance Matrix

**Covariance matrix**,  $\Sigma_{jk} = Cov(X_j, X_k) = E[(X_j - E[X_j])(X_k - E[X_k])] = \sigma_{jk}$ , is a square matrix

- The inverse of the covariance matrix is called "precision matrix",  $\Omega$ .
- $\Sigma$  is symmetric. (i.e.,  $Cov(X_i, X_j) = Cov(X_j, X_i)$ )
- $\Sigma$  is positive semi-definite (PSD).

$$u^T \Sigma u = \sum_{i,j=1}^n u_i \Sigma_{ij} u_j = \sum_{i,j=1}^n Cov(u_i X_i, u_j X_j) = Cov\left(\sum_i u_i X_i, \sum_j u_j X_j\right) \geq 0$$

- $\Sigma$ 's eigenvalues are non-negative because  $\Sigma$  is PSD.

## 3 Code

- For the setting, the numbers of observations in first cell, second cell, and third cell are 150, 150, 200.
- The number of features are 8.

## 1. First try

- First Cell Type  
Mean: 0, Cov: 0.1 but [(1, 3), (1, 4), (5, 6)]: 0.7, f:  $1.2 * \text{sign}(x_i) * \text{abs}(x_i)^2$
- Second Cell Type Mean: 0, Cov: 0.3 but [(1, 3), (1, 4)]: 0.7, [(5,7), (5,8)]: 0.6, [c(2,3), c(6,7)]: 0.4, f:  $1.5 * \text{sign}(x_i) * \text{abs}(x_i)$ .
- Third Cell Type Mean: 0, Cov: 0.5 [(5,6)]: 0.2, [(4,7)]: 0.3, f:  $0.8 * \text{sign}(x_i) * \text{abs}(x_i)^{0.8}$

